





Computational Ecosystem in Neuroscience:
High Performance Analytics and Computing for the HBP

- HBP is a Future and Emerging Technologies (FET) Flagship project co-funded by the European Union with a total duration of 10 years and up to 1 billion € funding
- > 100 partner institutions from 20 European countries (and others)

### Goals for 10 year endeavor

- Build a European Research Infrastructure (RI) for the neuroscience community. including six platforms:
  - Neuroinformatics, Brain Simulation, High Performance Analytics and Computing, Medical Informatics, Neuromorphic Computing, Neurorobotics
- Gather, organize and disseminate data describing brain and its diseases
- Construct large scale brain simulation
- Build multi-scale scaffold theory and models of the brain
- Develop brain-inspired computing, data analytics and robotics.

# **High Performance Computing and Analytic Platform**





We build and operate a supercomputing, data and visualization infrastructure enabling scientists to

- Run large-scale, data intensive, interactive brain simulations up to the size of a full human brain
- Manage the large amounts of data used and produced in the HBP
- Manage complex workflows comprising concurrent simulation, data analysis and visualization workloads













### Federated data & compute infrastructure



Aim	Providing scalable compute and data services in a federated manner
Target communities	<ul> <li>Neuroscience: HBP is prioritized driver for infrastructure design</li> <li>Materials science</li> <li>Genomics</li> <li>Physical science experiments</li> <li> and others with similar requirements</li> </ul>
Key features	<ul> <li>Scalable compute resources</li> <li>Federated data infrastructure</li> <li>Interactive Compute Services providing access to the federated data infrastructure as well as elastic access to the scalable compute resources</li> </ul>
Consortium	











### Software stacks



#### Inputs needing analysis:

- Electrophysiological data analysis
- fMRI analysis pipelines
- EEG data
- Microscopic data for brain atlases

#### Simulations for model testing:

- Neural mass models: 700 frequency coupled regions for a full brain
- Point neuron models: up to 10<sup>11</sup> neurons approximated as a network of time-delayed binary signals and an integration function defining firing times
- Morphologically detailed neuron models: neurons are modeled as combinations of synaptic functions, coupled passive cables with active variable resistors
- ... and at the bottom, reaction-diffusion models for synapses, signal exchanges and ion channels
- Multiscale models combining fast, coarse-grain models for global behavior and fine-grain models for perturbation-sensitive regions

#### Large scale analysis & machine learning

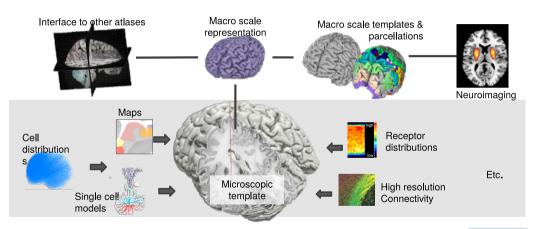
- Workflow construction toolset: Modular Science
- Parameter optimization tools: Learning-2-Learn, JuPeX

#### Visualization:

- Analytic pipelines visualization
- Analysis for network construction
- Interactive and steered simulations
- Experiment & provenance tracking

### **Human Brain Atlas**

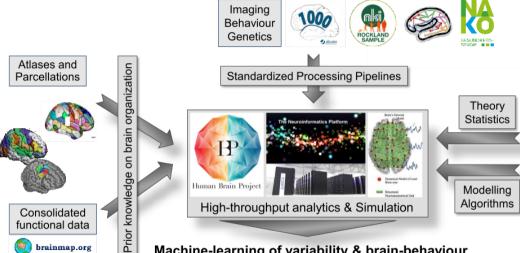




Timo Dickscheid

## **External data to analysis**





Machine-learning of variability & brain-behaviour relations in health and disease

Simon Eickhoff

brainmap.org

ANIMA

Sandra Diaz Pier

Deco, Jirsa, McIntosh Nat Rev Neurosci 2011 Deco, Jirsa Journ Neurosci 2012

Deco, Jirsa, McIntosh TINS 2013

Ritter et al Brain Connectivity 2013

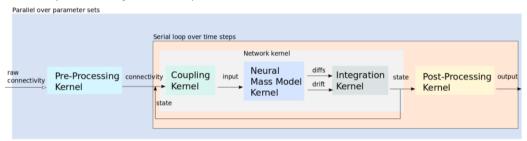
0 0.2 0.4 0.6 0.8 1

FC model

### TVB for HPC



"The Virtual Brain" software package for solving these systems is a combination of Matlab and Python scripts attached to a front-end visual control center, all which are designed for tradition desktops. To do parameter searches over individual dynamics for large data sets requires HPC resources



#### Approach:

Sandra Diaz Pier, Marmaduke Woodman (Blundell et al 2018)

- Automatic kernel code generation using 'loo.pv'
- Domain Specific Language to describe neural mass models, coupling, integration, and pre- and post-processing kernels.
- Use one description for multiple backends: CPUs, GPUs, KNL, FPGAs, ...

Collaboration w/ Institut des Neuroscience des Systèmes, Aix-Marseille Université



# NEST = NEural Simulation Tool Simulations of large networks of spiking point-neuron models (internal single-neuron dynamics can be arbitrarily complex)

- C++ kernel
- built-in simulation language interpreter (SLI)
- Python-based user interface (PyNEST)
- NESTml neuron modeling DSL
- multi-threading to use multi-processor machines efficiently
- MPI & OpenMP parallelism to use compute clusters and supercomputers
- Synaptic plasticity
- Structural plasticity
- Electrically coupled neurons

# **△**arbor: Background



### Arbor is a project to develop:

- a new morphologically detailed, multi-compartmental neuronal network simulator library,
- that is optimized for HPC systems,
- and is easy to integrate into existing workflows.

### Developed

- between the Jülich Supercomputing Centre at the Forschungszentrum Jülich and CSCS the Swiss Supercomputing Centre as a component of the Human Brain Project
- as an open-source project
- using an open developmental model

### See current development at

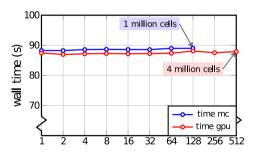
https://github.com/arbor-sim/arbor

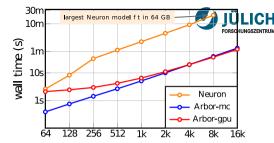
Hardware Software stacks Software Infrastructure I ...

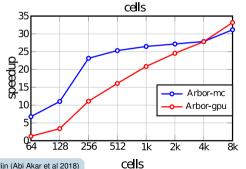
Crav XC40: 2x 18-core Broadwell per node Cray XC50: 1x P100 GPU per node 150 compartments & 10,000 synapses per cell

Passive dendrites, Hodgkin-Huxley soma ring network

network duration 100 ms





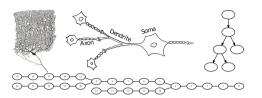


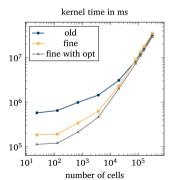
Ben Cumming, Sam Yates, Nora Abi Akar, Anne Küsters, Wouter Kliin (Abi Akar et al 2018)

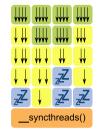
Detailed Workflow

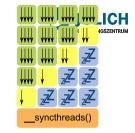
Arbor

# **△**arbor: GPU optimization

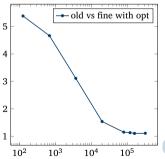












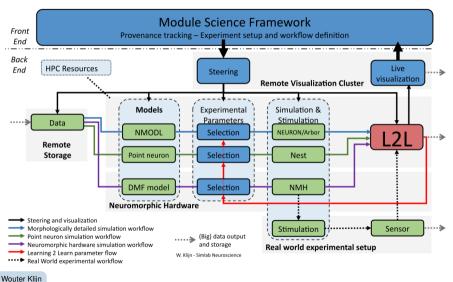
Felix Huber 2018

number of cells

Hardware Software stacks Software Infrastructure I ...

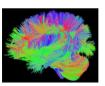
### **Modular Workflows**





### **Visualization Framework**











Caspers et al. 2014

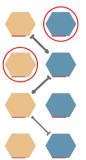
Pastor et. al. 2015, Data: BBP

Potjans et. al. 2011

Woodman, et al. 2017



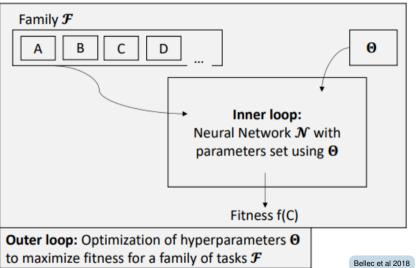




MSC Visualization 1.21

# Learning2Learn Concept





# **Workshops**



#### Symposium on generative connectomics and plasticity

Date: March 26-29 2019 Location: Jülich Germany

Join us in looking at the state of the art in simulation and modeling of generative connectomics and plasticity. We will discuss modeling and simulation of connectivity generation from two perspectives:

- neural development and structural plasticity in biological neural networks
- generation of connectivity in spiking and artificial neural networks

#### **HPAC** training

Date: December 11 & 12 Location: Barcelona

Training in HPAC infrastructure tools and a quick introduction to NEST & Arbor

#### Proposed: High Performance Computing for Neuroscience: Hands-on introduction to supercomputing usage, tools and applications

Date: June ?? (3 day course) Location: Jülich Germany

This workshop will cover the basics of supercomputing needed for getting started using HPC systems for (neuroscience) research, Including: Python, machine usage, data management, introduction to NEST, Arbor and visualization, and usage of other HBP platform tools such as neurorobotics software and neuromorphic hardware in the context of HPC.

### SimLab Neuroscience



## We're hiring

Are you a computer scientist or mathematician interested in all aspects of computational neuroscience at the intersection of high-performing computing?

Can you solve inverse problems, program GPUs or love to build systems with exotic neuromorphic hardware or build complex neuroscientific multiscale workflows?

The SimLab Neuroscience at the Jülich Supercomputing Center in the Forschungszentrum Jülich is hiring multiple positions.

Please contact a.peyser@fz-juelich.de.



### And thanks to...

#### SimLab

Sandra Diaz Piers Wouter Kliin Guido Trensch Jochen Eppler Lekshmi Deepu Anna Lührs Meredith Peyser Abigail Morrison Boris Orth **Bastian Tweddell** Kai Kraisek Anne Küsters Anne Carstens Andreas Herten

...and everyone else in the SimLab and HPCNS

#### Arbor: CSCS

Ben Cumming Sam Yates Nora Abi Akar Marco Kabic Vasileos Karakakis

#### **TVB** external

Viktor Jirsa, Marseille Marmaduke Woodhouse, Marseille Petra Ritter, Charité Michael Schirner, Charité Olaf Sporns, Indiana University

... and everyone else across the HBP

# Imaging: Institute for Neuroscience & Medicine

Katrin Amunts
Timo Dickscheid

Simon Eickhoff Thanos Manos

Felix Hoffstaedter Popovych, Oleksandr

#### **NEST: NMBU**

Susanne Kunkel Hans Ekkehard Plesser

#### Graz Technische Universität

Wolfgang Maas Anand Subramoney

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daint-mc | Cray XC40: 2× 18-core Broadwell per node

daint-gpu Cray XC50: 1×P100 GPU per node

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